Processing Raw Text POS Tagging

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Outline

- Accessing Text beyond NLTK
- Processing Raw Text
- POS Tagging

NLTK includes a good selection of various corpora among which a small selection of texts from the Project Gutenberg electronic text archive. Project Gutenberg contains more than 50 000 free electronic books, hosted at http://www.gutenberg.org.



Unfortunately, only 18 books are provided, which you can list as we have seen before:

Accessing the original collection is thus helpful:

```
import nltk
import urllib
url="http://www.gutenberg.org/files/2554/2554-0.txt"
urlData = urllib.request.urlopen(url)
firstLine = urlData.readline().decode("utf-8")
print(firstLine)
# prints
# The Project Gutenberg EBook of Crime and Punishment
    , by Fyodor Dostoevsky
```

2554 = Crime and Punishment, by Fyodor Dostoevsky

How do I find out the Book IDs?



Directly from the Gutenberg Webpage \rightarrow not very efficient.



Under the link http://www.gutenberg.org/dirs/GUTINDEX.ALL, the project provides a file listing all available books and their IDs.

```
**** The Language of the eBooks is English, unless otherwise noted ****
~ ~ ~ Posting Dates for the below eBooks: 1 Nov 2013 to 30 Nov 2013 ~ ~ ~ ~
TITLE and AUTHOR
                                                                      ETEXT NO.
Ypres 1914, by Otto Schwink
                                                                          44234
 [Subtitle: An Official Account Published by Order
  of the German General Staff1
 [Translator: Graeme Chamley Wynne]
Plays by August Strindberg, Third Series, by August Strindberg
                                                                          44233
  [Translator: Edwin Bj?rkman]
L'Illustration, No. 1608, 20 d?cembre 1873, by Various
                                                                          44232
  [Language: French]
The Miraculous Medal, by Jean Marie Aladel
                                                                          44231
 [Subtitle: Its Origin, History, Circulation, Results]
```



And as you can see Crime and Punishment, by Fyodor Dostoevsky is also listed there:

Under the Redwoods, by Bret Harte Contents: Jimmy's Big Brother From California The Youngest Miss Piper A Widow Of The Santa Ana Valley The Mermaid Of Lighthouse Point Under The Eaves How Reuben Allen "Saw Life" In San Francisco Three Vagabonds Of Trinidad A Vision Of The Fountain A Romance Of The Line Bohemian Days In San Francisco Under The Redwoods	2555
Crime and Punishment, by Fyodor Dostoevsky	2554
[Tr.: Constance Garnett]	
Jeanne d'Arc, by Mrs. (Margaret) Oliphant	2553
[Subtitle: Her Life And Death]	
Thankful's Inheritance, by Joseph C. Lincoln	2552
Droll Stories, Volume 3, by Honore de Balzac	2551
{See also: #2318 & #1925)	

So what more can we find out?



Regex to extract information

- Recall: Start of String and End of String Anchors:
 - matches the position before the first character in the string \rightarrow Applying a to abc matches a.
 - ullet Similarly, \ullet matches right after the last character in the string ullet
 - c \$ matches c in abc
- Check regex via https://regex101.com/

Regex to extract information

- Lookahead and lookbehind, collectively called "lookaround",
 are zero-length assertions just like start and end of word anchors
- They do not consume characters in the string, but only assert whether a match is possible or not
- **Positive Lookbehind**: $(? \le B)A \rightarrow \text{find expression A where expression B precedes:}$
- Positive Lookahead: A(? = B) → find expression A where expression B follows
- Negative Lookbehind: (? <!B)A → Find expression A where expression B does not precede
- Negative Lookahead: A(?!B) → Find expression A where expression B does not follow



How to extract "French" from "[Language: French]" ???

o Check regex via https://regex101.com/

About 109 languages considering mixtures, such as *German to English*:

```
import re
f = open("GUTINDEX.ALL", encoding="utf-8", errors="ignore")
data = f.read()
s = set(re.findall("(?<=\[Language: ).*?(?=\s*?\])", data))
print(s)
#{"German and English", "Bulgarian", "Friulian", "Spanish, English and Tagalog", "Catalan", "Ojibwa", "German to English", "Welsh", "French and English", "Hebrew", "Russian", "Spanish", "Galician", "Greek", "Spanish and Tagalog", "Romani, dialecto de los Gitanos de Espaa", "Portuguese & French", ...
}</pre>
```

findall() returns all non-overlapping matches of pattern in string as a **list of strings**.

```
data = open("GUTINDEX.ALL", encoding="utf-8", errors="ignore").
    read()

dict={}

for m in re.finditer("(?<=\n)(.*?)\s+(\d+)(?=\n)", data):

    dict[m.group(2)]= m.group(1)

print(dict)

#{"9333": "Johnny Bear, by E. T. Seton", "11513": "On Land And Sea
    At The Dardanelles, by Thomas Charles Bridges", "12461": "
    Castles in the Air, by Baroness Emmuska Orczy", "11034": "A
    Compilation of the Messages and Papers of the Presidents,
    Richardson", "3014": "The Old Northwest, by Frederic Austin
Ogg", ... }</pre>
```

finditer() returns iterator yielding match objects (use it if the number of matches is really high).



Dealing with other formats

Often enough, content on the Internet as well as locally stored content is transformed to a number of formats different from plain text (.txt).

- RTF Rich Text Format (.rtf)
- HTML HyperText Markup Language (.html, .htm)
- XHTML Extensible HyperText Markup Language (.xhtml, .xht, .xml, .html, .htm)
- XML Extensible Markup Language (.xml)
- RSS Rich Site Summary (.rss, .xml)

Dealing with other formats

Additionally, often text is stored in binary formats, such as:

- MS Office formats (.doc, .dot, .docx, .docm, .dotx, .dotm, .xls, .xlt, .xlm, .ppt, .pps, .pptx ... and many others)
- PDF Portable Document Format (.pdf)
- OpenOffice formats (.odt, .ott, .oth, .odm ...
 and others)

HTML

http:

//www.bbc.com/news/world-middle-east-42412729

```
import urllib
url="http://www.bbc.com/news/world-middle-east-42412729"
urlData = urllib .request.urlopen(url)
html = urlData.read().decode("utf-8")
print (html)
# prints
#'<!DOCTYPE html>\n<html lang="en" id="responsive-news">\n
#<head prefix="og: http://ogp.me/ns#">\n <meta charset="utf-8"
    ">|n|
# <meta http-equiv="X-UA-Compatible" content="IE=edge,chrome=1">\n
# <title >Yemen rebel ballistic missile \'intercepted over Riyadh\'
     - BBC News</title >\n
```

HTML

HTML is often helpful since it marks up the distinct parts of the document, which makes them easy to find:

- Python library for pulling data out of HTML and XML files.
- can navigate, search, and modify the parse tree.

```
html doc =
<html><head><title >The Dormouse's story </title ></head>
<body>
<b>The Dormouse's story </b>
Once upon a time there were three little sisters;
     and their names were
<a href="http://example.com/elsie" class="sister" id="link1">Elsie
    </a>.
<a href="http://example.com/lacie" class="sister" id="link2">Lacie
    </a> and
<a href="http://example.com/tillie" class="sister" id="link3">
    Tillie </a>:
and they lived at the bottom of a well. 
 ...
```

```
from bs4 import BeautifulSoup
soup = BeautifulSoup(html_doc, 'html.parser')

#with open("index.html") as fp:
soup = BeautifulSoup(fp)
```

BeautifulSoup object represents the document as a nested data structure:

```
from bs4 import BeautifulSoup
   soup = BeautifulSoup(html_doc, 'html.parser')
   print(soup.prettify())
   # <html>
   # <head>
   # <title >
   # The Dormouse's story
   # </title >
   # </head>
   # <body>
     <h>>
       The Dormouse's story
14
        </h>
```

Simple ways to navigate that data structure: say the name of the tag you want

```
soup.title
  # <title >The Dormouse's story </title >
4
  soup.title.string
  # u'The Dormouse's story'
  soup.title.parent.name
  # u'head'
  soup.p
  # <b>The Dormouse's story </b> 
  soup.p['class']
   # u'title '
```

Simple ways to navigate that data structure:

```
soup.a
   # <a class="sister" href="http://example.com/elsie" id="link1">
        Elsie </a>
4
   soup.find_all('a')
   # [<a class="sister" href="http://example.com/elsie" id="link1">
        Elsie </a>.
   # <a class="sister" href="http://example.com/lacie" id="link2">
       Lacie </a>.
   # <a class="sister" href="http://example.com/tillie" id="link3">
        Tillie </a>1
   soup.find(id="link3")
   # <a class="sister" href="http://example.com/tillie" id="link3">
        Tillie </a>
```

One common task is extracting all the URLs found within a page's <a> tags:

```
1 for link in soup.find_all('a'):
2    print(link.get('href'))
3 # http://example.com/elsie
4 # http://example.com/lacie
5 # http://example.com/tillie
```

Another common task is extracting all the text from a page:

```
print(soup.get text())
# The Dormouse's story
#
# The Dormouse's story
#
# Once upon a time there were three little sisters;
   and their names were
# Elsie,
# Lacie and
# Tillie:
# and they lived at the bottom of a well.
#
```

Installing Beautiful Soup:

apt-get install python3-bs4 (for Python 3)

Nowadays we often store text in formats that are not human-readable: e.g. binary format (e.g. .doc, .pdf). These formats are not as easily processed as simple text.

There are a number of third-party modules that can be installed and used for extracting data from binary files. Yet, depending on the files, the output is not always clean and easily usable.

```
import nltk
import PyPDF2

pdf = PyPDF2.PdfFileReader(open("text.pdf", "rb"))

for page in pdf.pages:
    print(page.extractText())

# prints each of the pages from as raw text.
```

Snippet from a pdf document "intro.pdf"



Symbolische Programmiersprache

Abstract This course will use the Python programming language as the basis for various computational linguistic implementations. We will cover a wide range of natural language processing (NLP) tasks, such as tokenization, keyword extraction, normalization and stemming, categorization and tagging, as well as classification, chunking and language identification. All the latter are basic NLP tasks that will be discussed and their implementation in Python will be realized during the practical exercise in connection to the course. With respect to each task, we will concentrate on the problems that this task faces and the possible solutions to them within the Python framework. All students will be required to complete weekly assignments and write a term paper (10-12 pages) as a summary of the discussed topics and their importance and application for computational linguistics.



```
import nltk
import PyPDF2

pdf = PyPDF2.PdfFileReader(open("intro.pdf", "rb"))

for page in pdf.pages:
    print(page.extractText()+"\n")
```

In other cases, the full text might be extracted, but not in a easily usable format as here:

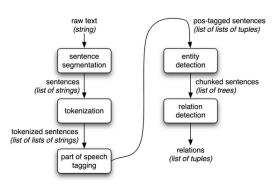
SymbolischeProgrammierspracheAbstractThiscoursewillusethePythonprogramminglanguagea sthebasisforvariouscomputationallinguisticimplementations. Wewillcoverawiderangeofna turallanguageprocessing (NLP) tasks, suchastokenization, keywordextraction, normalizatio nandstemming, categorizationandtagging, aswellaschunkingandlanguageidenAllthelatterar ebasicNLFtasksthatwillbediscussedandtheirimplementationinPythonwillberealizedduring thepracticalexerciseinconnectiontothecourse. Withrespecttoeachtask, wewillconcentrate ontheproblemsthatthistaskfacesandthepossiblesolutionstothemwithinthePythonframework. Allstudentswillberequiredtocompleteweeklyassignmentsandwriteatermpaper (10-12pages) asasummaryofthediscussedtopicsandtheirimportanceandapplicationforcomputationallinguistics. FormatofthecourseCredits: 4SW8 (6ECTS) Coursetimes: Tuesdays16:00(18:00, Thursdays16:00(18:00, Thursdays16

What next?

We have seen multiple ways of getting raw text? But what to do with it next?



NLP pipelines



Tokenization

Tokenization with NLTK:

```
import nltk
import urllib

url="http://www.gutenberg.org/files/2554/2554-0.txt"

urlData = urllib.request.urlopen(url)
data = urlData.read().decode("utf-8")

tokens = nltk.word_tokenize(data)

print(tokens)
```

Tokenization

Tokenization with regex:

```
url="http://www.gutenberg.org/files/2554/2554—0.txt"
  urlData = urllib.request.urlopen(url)
  data = urlData.read().decode("utf-8")
4
  for m in re.finditer("\w+", data):
       print (m. group () )
6
  # prints:
  # ...
  # him
  # although
  # the
  # explanation
```

Tokenization

```
url="http://www.gutenberg.org/files/2554/2554-0.txt"
urlData = urllib.request.urlopen(url)
data = urlData.read().decode("utf-8")

for m in re.finditer("\w+", data):
    # for m in re.finditer("\S+", data):
    # for m in re.finditer("[a-zA-Z]+", data):
    # ...

print(m.group())
```

Searching Tokenized Text

```
import nltk
  from nltk.corpus import gutenberg
4
  moby = nltk.Text(gutenberg.words("melville-moby_dick.txt"))
  print(moby.findall("<a> <.*> <man>"))
  # prints
  # a monied man; a nervous man; a dangerous man; a white man
       ; a white man; a white man; a pious man; a queer man;
      a good man; a mature man; a white man; a Cape man; a
       great man; a wise man; a wise man; a butterless man; a
        white man; a fiendish man; a pale man; a furious man;
       a better man; a certain man; a complete man;
```

Searching Tokenized Text

- It is easy to build search patterns when the linguistic phenomenon we're studying is tied to particular words.
- For instance, searching a large text corpus for expressions of the form x and other ys allows us to discover hypernyms.

Searching Tokenized Text

import nltk

```
from nltk.corpus import brown
4
  hobbies_learned = nltk.Text(brown.words(categories=["
      hobbies", "learned"]))
  print(hobbies learned.findall(r"<\w+> <and> <other> <\w+s>"
      ))
  # prints
  # speed and other activities; water and other liquids; tomb
       and other landmarks: Statues and other monuments:
       pearls and other jewels; charts and other items; roads
       and other features; figures and other objects;
       military and other areas: demands and other factors:
       abstracts and other compilations; iron and other
       metals
```

POS Tagging Overview Documentation Disambiguation Example from Brown Variation across Tagsets Tagged Corpora for Other Languages Frequency Distributions of POS Tags Example Explorations

POS Tagging Overview

- parts-of-speech (word classes, lexical categories, POS) e.g. verbs, nouns, adjectives, etc.
- part-of-speech tagging (POS tagging, tagging) labeling words according to their POS
- tagset the collection of tags used for a particular task

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Using a Tagger

A part-of-speech tagger, or POS tagger, processes a sequence of words, and attaches a part of speech tag to each word:

```
import nltk

text = nltk.word_tokenize("And now for something completely different")
print(nltk.pos_tag(text))

# [('And', 'CC'), ('now', 'RB'), ('for', 'IN'), ('something', 'NN'), ('completely', 'RB'), ('different', 'JJ')]
```

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Variation in Tags

```
1 # [('And', 'CC'), ('now', 'RB'), ('for', 'IN'), ('something', 'NN'), ('completely', 'RB'), ('different', 'JJ')]
```

- CC coordinating conjunction
- RB adverb
- IN preposition
- NN noun
- JJ adjective



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Documentation

NLTK provides documentation for each tag, which can be queried using the tag, e.g:

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Documentation

Note!

Some POS tags denote variation of the same word type, e.g. NN, NNS, NNP, NNPS, such can be looked up via regular expressions.

```
    >>> nltk.help.upenn_tagset('NN*')
    NN: noun, common, singular or mass
    common-carrier cabbage knuckle-duster Casino ...
    NNP: noun, proper, singular
    Motown Venneboerger Czestochwa Ranzer Conchita ...
    NNPS: noun, proper, plural
    Americans Americas Amharas Amityvilles ...
    NNS: noun, common, plural
    undergraduates scotches bric-a-brac ...
```

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Disambiguation

POS tagging does not always provide the same label for a given word, but decides on the correct label for the specific context – disambiguates across the word classes.

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Disambiguation

POS tagging does not always provide the same label for a given word, but decides on the correct label for the specific context – disambiguates across the word classes.

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Example from Brown

Whenever a corpus contains tagged text, the NLTK corpus interface will have a tagged_words() method.

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Variation across Tagsets

Even for one language, POS tagsets may differ considerably!



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Variation across Tagsets

Alphabetical list of part-of-speech tags used in the Penn Treebank Project:

Number	Tag	Description	
1.	CC	Coordinating conjunction	
2.	CD	Cardinal number	
3.	DT	Determiner	
4.	EX	Existential there	
5.	FW	Foreign word	
6.	IN	Preposition or subordinating conjunction	
7.	JJ	Adjective	
8.	JJR	Adjective, comparative	
9.	JJS	Adjective, superlative	
10.	LS	List item marker	
11.	MD	Modal	
12	NN	Noun singular or mass	

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Variation across Tagsets

The Open Xerox English POS tagset:

no opoi	The open Kerek English Fee tageet.				
Tag	Description	Example			
+ADJ	(basic) adjective	[a] blue [book], [he is] big			
+ADJCMP	comparative adjective	[he is] bigger, [a] better [question]			
+ADJING	adjectival ing-form	[the] working [men]			
+ADJPAP	adjectival past participle	[a] locked [door]			
+ADJPRON	pronoun (with determiner) or adjective	[the] same; [the] other [way]			
+ADJSUP	superlative adjective	[he is the] biggest; [the] best [cake]			
+ADV	(basic) adverb	today, quickly			
+ADVCMP	comparative adverb	sooner			
+ADVSUP	superlative adverb	soonest			
+CARD	cardinal (except one)	two, 123, IV			
+CARDONE	cardinal one	[at] one [time]; one [dollar]			
+CM	comma	1			
+COADV	coordination adverbs either, neither	either [by law o by force]; [he didn't come] either			
+COORD	coordinating conjunction	and, or			



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Variation across Tagsets

The variation across tagsets is based on the different decisions and the information needed to be included:

- morphologically rich tags
- morphologically poor ones

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Arabic Example

For example, in Arabic the morphologically-poor tag $\mathbb{N}\mathbb{N}$ may be divided into the following morphologically-rich variants:

```
(ABBREV NN)
(LATIN NN)
                                        (NOUN+CASE DEF ACC NN)
                                        (NOUN+CASE DEF GEN NN)
(DET+NOUN NN)
                                        (NOUN+CASE DEF NOM NN)
(DET+NOUN+NSUFF FEM SG NN)
(NOUN NN)
                                        (NOUN+CASE INDEF ACC NN)
(NOUN+NSUFF_FEM_SG NN)
                                        (NOUN+CASE INDEF GEN NN)
                                        (NOUN+CASE INDEF NOM NN)
(NOUN+NSUFF MASC SG ACC INDEF NN)
(DEM+NOUN NN)
                                        (NOUN+NSUFF FEM SG+CASE DEF ACC NN)
(DET+NOUN+CASE DEF ACC NN)
                                        (NOUN+NSUFF FEM SG+CASE DEF GEN NN)
(DET+NOUN+CASE DEF GEN NN)
                                        (NOUN+NSUFF FEM SG+CASE DEF NOM NN)
(DET+NOUN+CASE DEF NOM NN)
                                        (NOUN+NSUFF FEM SG+CASE INDEF ACC NN)
(DET+NOUN+NSUFF FEM SG+CASE DEF ACC NN) (NOUN+NSUFF FEM SG+CASE INDEF GEN NN)
(DET+NOUN+NSUFF FEM SG+CASE DEF GEN NN) (NOUN+NSUFF FEM SG+CASE INDEF NOM NN)
(DET+NOUN+NSUFF FEM SG+CASE DEF NOM NN) (NEG PART+NOUN NN)
```

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NLTK and simplified tags

NLTK includes built-in mapping to a simplified tagset for most complex tagsets included in it:

```
1 >>> nltk.corpus.brown.tagged_words()
2 [('The', 'AT'), ('Fulton', 'NP-TL'), ...]
3
4 >>> nltk.corpus.brown.tagged_words(tagset='universal')
5 [('The', 'DET'), ('Fulton', 'NOUN'), ...]
```

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NLTK and simplified tags

The Universal Part-of-Speech Tagset of NLTK:

Tag	Meaning	English Examples
ADJ	adjective	new, good, high, special, big, local
ADP	adposition	on, of, at, with, by, into, under
ADV	adverb	really, already, still, early, now
CONJ	conjunction	and, or, but, if, while, although
DET	determiner, article	the, a, some, most, every, no, which
NOUN	noun	year, home, costs, time, Africa
NUM	numeral	twenty-four, fourth, 1991, 14:24
PRT	particle	at, on, out, over per, that, up, with
PRON	pronoun	he, their, her, its, my, I, us
VERB	verb	is, say, told, given, playing, would
	punctuation marks	.,;!
Х	other	ersatz, esprit, dunno, gr8, univeristy

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Tagged Corpora for Other Languages

Tagged corpora for several other languages are distributed with NLTK, including Chinese, Hindi, Portuguese, Spanish, Dutch, and Catalan.

```
1 >>> nltk.corpus.sinica_treebank.tagged_words()
2 >>> nltk.corpus.indian.tagged_words()
```

```
[('一', 'Neu'), ('友情', 'Nad'), ('嘉珍', 'Nba'), ...]
[('মহিবের', 'NN'), ('শন্তান', 'NN'), (':', 'SYM'), ...]
```

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Frequency Distributions of POS Tags

We have calculated Frequency Distributions based on a sequence of words. Thus, we can do so for POS tags as well.

```
import nltk
  from nltk.corpus import brown
4
  brown news tagged = brown.tagged words(categories='news',
      tagset='universal')
  tag fd = nltk.FreqDist(tag for (word, tag) in
      brown news tagged)
  print(tag_fd.most_common())
  # [('NOUN', 30640), ('VERB', 14399), ('ADP', 12355), ('.',
       11928), ('DET', 11389), ('ADJ', 6706), ('ADV', 3349),
       ('CONJ', 2717), ('PRON', 2535), ('PRT', 2264), ('NUM',
       2166), ('X', 106)]
```

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Example Explorations

```
import nltk
wsj = nltk.corpus.treebank.tagged_words(tagset='universal')
cfd1 = nltk.ConditionalFreqDist(wsj)
print(list(cfd1['yield'].keys()))
print(list(cfd1['cut'].keys()))
```

???

What is calculated in the lines 4 and 5?

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Example Explorations

```
import nltk
wsj = nltk.corpus.treebank.tagged_words(tagset='universal')
cfd1 = nltk.ConditionalFreqDist(wsj)
print(list(cfd1['yield'].keys()))
# ['NOUN', 'VERB']
print(list(cfd1['cut'].keys()))
// # ['NOUN', 'VERB']
```

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We can reverse the order of the pairs, so that the tags are the conditions, and the words are the events. Now we can see likely words for a given tag:

```
import nltk

wsj = nltk.corpus.treebank.tagged_words(tagset='universal')

cfd2 = nltk.ConditionalFreqDist((tag, word) for (word, tag)
    in wsj)

print(list(cfd2['VERB'].keys()))

# ['sue', 'leaving', 'discharge', 'posing', 'redistributing', 'emerges', 'anticipates', 'Hold', 'purrs', 'telling', 'obtained', 'ringing', 'mind', ...]
```

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???

What is calculated here?

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```
combined to achieve
# continue to place
# serve to protect
# wanted to wait
# allowed to place
# expected to become
# expected to approve
# expected to make
# intends to make
# seek to set
# like to see
# designed to provide
```

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???

What is calculated here?

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Extract most ambiguous words across the word classes:

```
import nltk
  from nltk.corpus import brown
4
  brown news tagged = brown tagged words(categories='news', tagset='
       universal')
  data = nltk.ConditionalFreqDist((word.lower(), tag) for (word, tag
       ) in brown news tagged)
  for word in data.conditions():
       if len(data[word]) > 3:
          tags = data[word].keys()
           print (word, ' '.join(tags))
  # that DET ADP ADV PRON
  # best ADJ NOUN ADV VERB
  # present ADJ NOUN ADV VERB
   # close NOUN ADJ ADV VERB
```

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TreeTagger

- The TreeTagger is a tool for annotating text with part-of-speech and lemma information
- used to tag German, English, French, Italian, Danish, Dutch, Spanish, Bulgarian, Russian, Portuguese, Galician, Greek, Chinese, Swahili, Slovak, Slovenian, Latin, Estonian, etc.

word	pos	lemma
The	DT	the
TreeTagger	NP	TreeTagger
is	VBZ	be
easy	JJ	easy
to	TO	to
use	VB	use

Sample output:

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Tree Tagoer

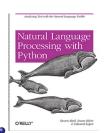
TreeTagger

- Download the files from http://www.cis. uni-muenchen.de/~schmid/tools/TreeTagger/
- Run the installation script: sh install-tagger.sh
- Test it:

```
echo 'Das ist ein gutes Beispiel!' | cmd/tree-tagger-german
        reading parameters
        tagging
          finished.
das
        PDS
                 die
ist
        VAFIN
                 sein
ein
        ART
                 eine
gutes
        AD.JA
                 gut
Beispiel
                 NN
                          Beispiel
        $.
```

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References



http://www.nltk.org/book/

https://github.com/nltk/nltk